
Mini Projet ML

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MPDS1

Partie 1 :

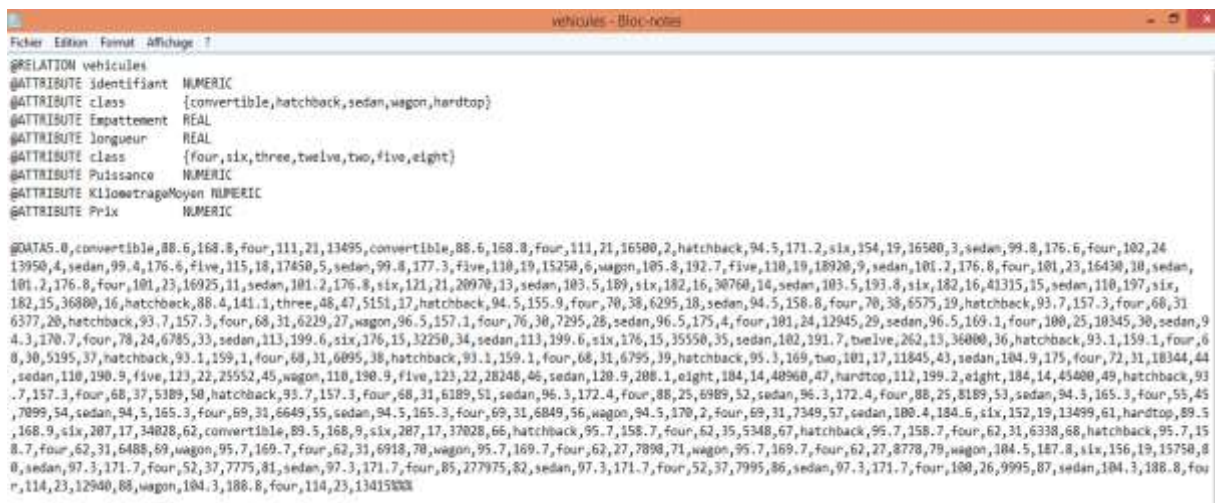
Question 1 : préparation du fichier Arff

Un fichier *arff* comprend toujours trois types d'informations : un nom pour la base de données, des attributs et des données. La chaîne de caractères @RELATION permet de donner un nom à la base de données. Par exemple, dans le cas du fichier *iris.arff*, le nom donné est *iris*.

@RELATION iris

Dans notre cas et selon le fichier « caracteristiquevehicules.csv » on a 8 attributs :

- Identifiant
- Carrosserie
- Empattement
- Longueur
- Nombre-cylindres
- Puissance
- KilometrageMoyen
- Prix



```
Fichier Edition Format Affichage ?
vehicles - Bloc-notes
@RELATION vehicles
@ATTRIBUTE identifiant NUMERIC
@ATTRIBUTE class {convertible,hatchback,sedan,wagon,hardtop}
@ATTRIBUTE empattement REAL
@ATTRIBUTE longueur REAL
@ATTRIBUTE class {four,six,three,twelve,two,five,eight}
@ATTRIBUTE puissance NUMERIC
@ATTRIBUTE kilometrageMoyen NUMERIC
@ATTRIBUTE prix NUMERIC

@DATA 0,convertible,88.6,168.8,four,111,21,13495,convertible,88.6,168.8,four,111,21,16500,2,hatchback,94.5,171.2,six,154,19,16500,3,sedan,99.8,176.6,four,102,24,13950,4,sedan,99.4,176.6,five,115,18,17450,5,sedan,99.8,177.3,five,110,19,15250,6,wagon,105.8,192.7,five,110,19,18920,9,sedan,101.2,176.8,four,101,23,16450,10,sedan,101.2,176.8,four,101,23,16925,11,sedan,101.2,176.8,six,121,21,20970,13,sedan,103.5,189,six,182,16,30760,14,sedan,103.5,193.8,six,182,16,41315,15,sedan,110,197,six,182,15,36800,16,hatchback,88.4,141.1,three,48,47,5151,17,hatchback,94.5,155.9,four,70,38,6295,18,sedan,94.5,158.8,four,70,38,6575,19,hatchback,93.7,157.3,four,68,31,6377,20,hatchback,93.7,157.3,four,68,31,6229,27,wagon,96.5,157.1,four,76,30,7295,28,sedan,96.5,175.4,four,101,24,12945,29,sedan,96.5,169.1,four,100,25,10345,30,sedan,94.3,170.7,four,78,24,6785,33,sedan,113,199.6,six,176,15,32250,34,sedan,113,199.6,six,176,15,35550,35,sedan,102,191.7,twelve,262,13,36000,36,hatchback,93.1,159.1,four,68,30,5195,37,hatchback,93.1,159.1,four,68,31,6095,38,hatchback,93.1,159.1,four,68,31,6795,39,hatchback,95.3,169,two,101,17,11845,43,sedan,104.9,175,four,72,31,18344,44,sedan,110,190.9,five,123,22,25552,45,wagon,110,190.9,five,123,22,28240,46,sedan,120.9,208.1,eight,184,14,40960,47,hardtop,112,199.2,eight,184,14,45400,49,hatchback,93.7,157.3,four,68,37,5309,50,hatchback,93.7,157.3,four,68,31,6189,51,sedan,96.3,172.4,four,88,25,6909,52,sedan,96.3,172.4,four,88,25,8189,53,sedan,94.5,165.3,four,55,45,7099,54,sedan,94.5,165.3,four,69,31,6649,55,sedan,94.5,165.3,four,69,31,6849,56,wagon,94.5,170.2,four,69,31,7349,57,sedan,100.4,184.6,six,152,19,13499,61,hardtop,89.5,168.9,six,207,17,34028,62,convertible,89.5,168.9,six,207,17,37028,66,hatchback,95.7,158.7,four,62,35,5348,67,hatchback,95.7,158.7,four,62,31,6338,68,hatchback,95.7,158.7,four,62,31,6488,69,wagon,95.7,169.7,four,62,31,6918,70,wagon,95.7,169.7,four,62,27,7898,71,wagon,95.7,169.7,four,62,27,8778,79,wagon,104.5,187.8,six,156,19,15750,80,sedan,97.3,171.7,four,52,37,7775,81,sedan,97.3,171.7,four,85,277975,82,sedan,97.3,171.7,four,52,37,7995,86,sedan,97.3,171.7,four,100,26,9995,87,sedan,104.3,188.8,four,114,23,12940,88,wagon,104.3,188.8,four,114,23,134150000
```

Figure 1: fichier arff

Après la préparation, on va importer le fichier .arff sur le weka.

Question 2 :

Dans cette partie on va planifier le prétraitement nécessaire sur les données avec la méthode de filtre non supervisé pour l'attribut prix comme montre la figure 2.

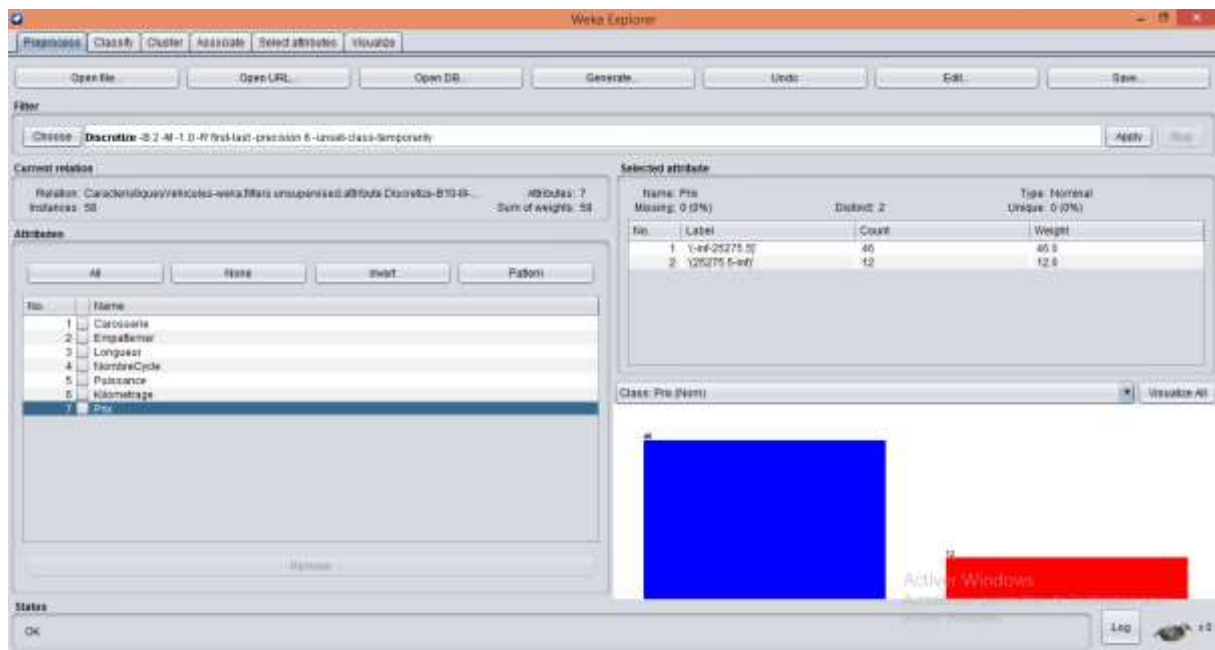


Figure 2: filtre discretize

Avec ce prétraitement on distingue deux classes ou cluster, bas (avec couleur rouge et un cout de 42) et élevé (avec couleur bleu avec un cout de 12) et chaque élément doit appartenir à un seul cluster.

Question 3 :

Maintenant, on va effectuer un clustering avec la méthode K-means, nous utiliserons cet algorithme pour regrouper à partir des 6 attributs ou caractéristique pour caractériser les prix de véhicules résultants. L'avantage de cet algorithme est de gérer automatiquement un mélange d'attributs catégoriques et numériques. De plus, l'algorithme normalise automatiquement les attributs numériques lors des calculs de distance.

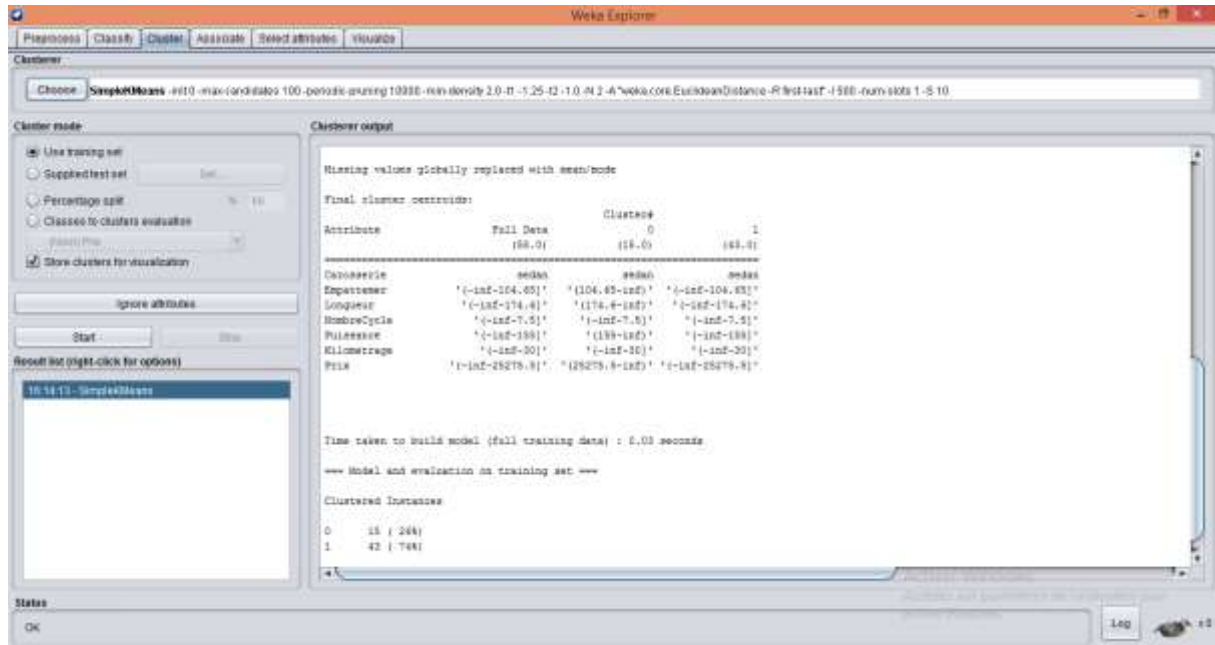


Figure 3: K-means

Après le clustering on peut identifier le deux cluster 0 et 1 avec leurs instances 15 et 41.

Question 4 :

La différence entre le partitionnement obtenu au départ et le partitionnement obtenu avec la méthode k-means consiste au quelques éléments qui appartient au premier cluster devient appartient au deuxième cluster.

Question 5 :

En appliquant la méthode «elbow method » avec un intervalle de nombre de groupes k entre 2 et 20 puis on va schématiser un courbe en fonction de cout dans l'axe de y et le nombre de clusters dans l'axe de x. l'objectif de cette technique est pour obtenir le cluster optimale de cet ensemble des données.

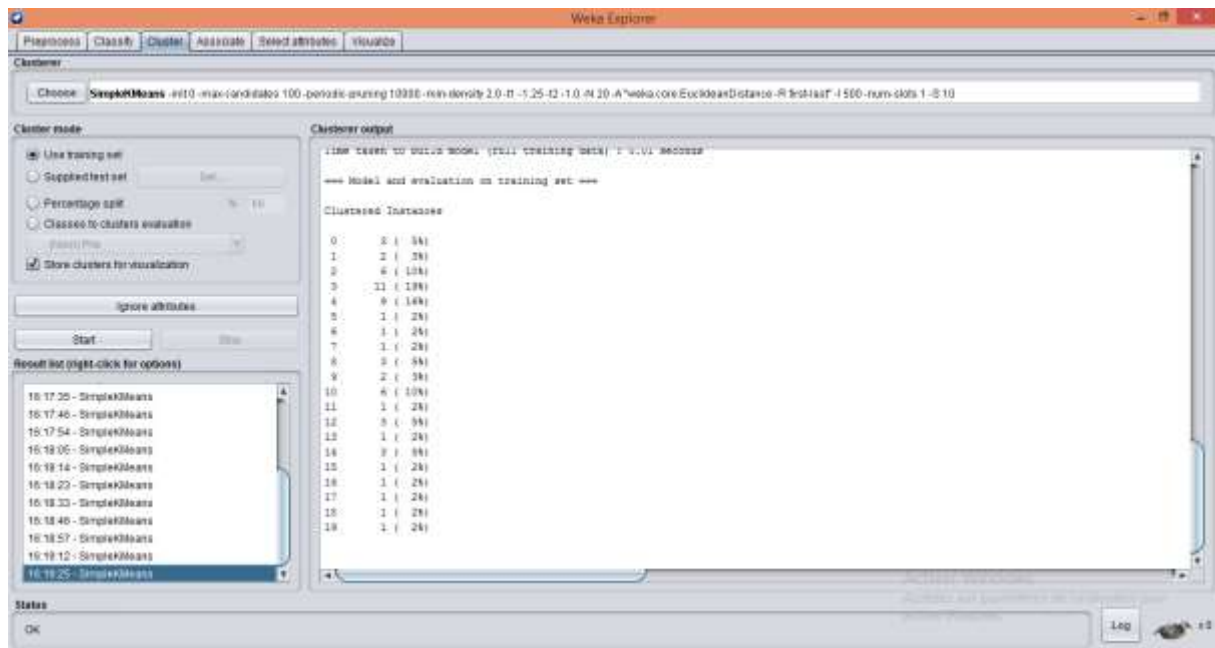


Figure 4: elbow method

La technique est très simple, chaque fois on va changer le nombre de cluster « incrémenter a 1 » et on restart l'algorithme pour obtenir le cout qui correspond à la sum of squared errors jusqu'à au nombre de cluster égal à 19. Ensuite on passe à l'étape de construction de schéma qui nous permet d'identifier le cluster optimal.

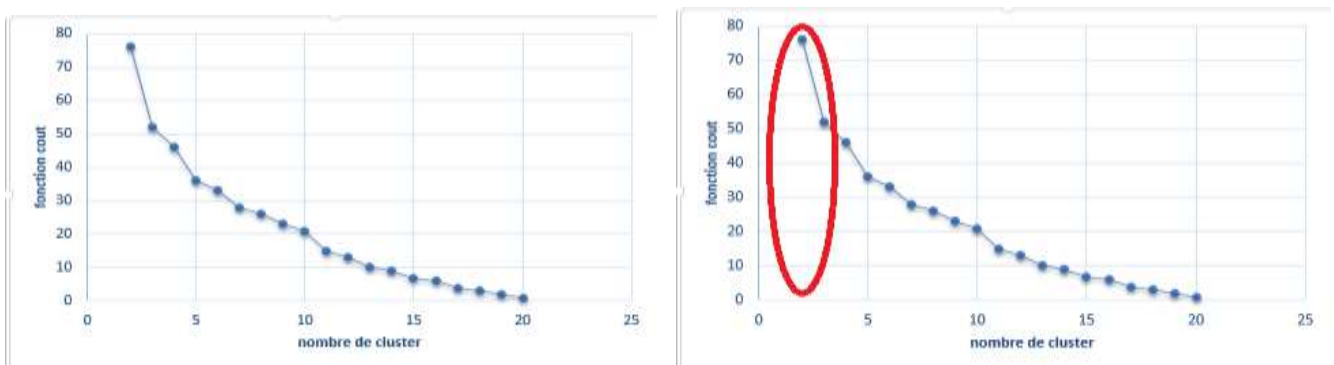


Figure 5: courbe « elbow method »

On peut conclure à partir de figure 5 que le cluster optimal est égale 3.

Question 6 :

Pour le cluster 3 la méthode K-means donne le résultat suivant

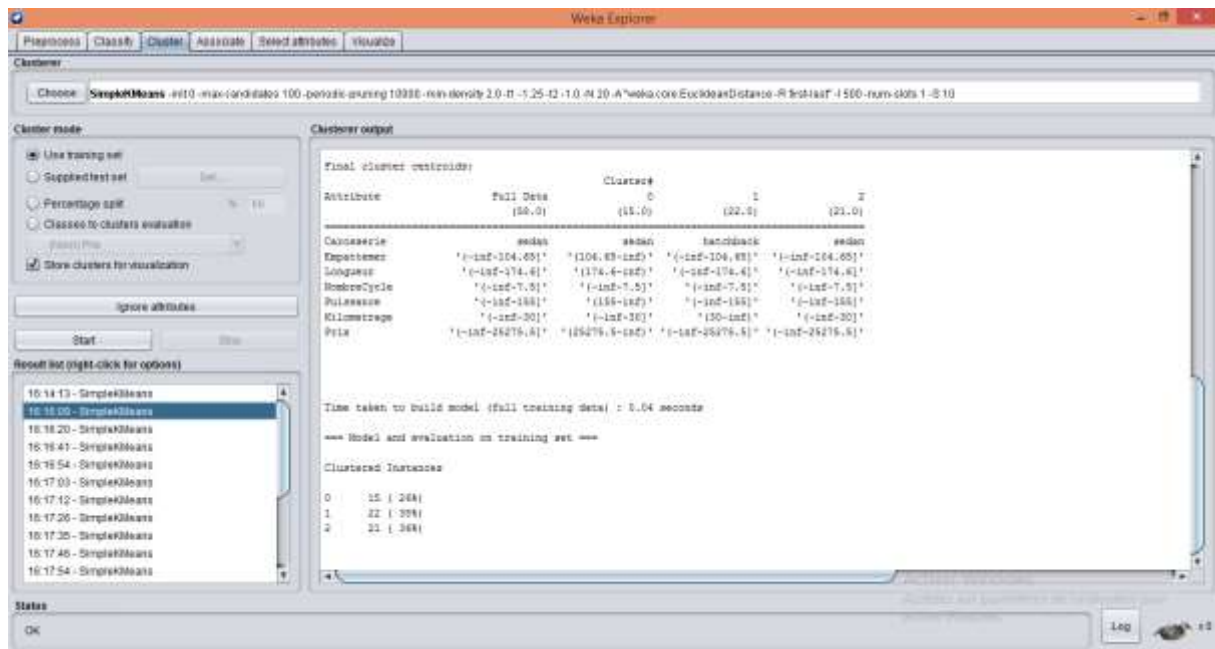


Figure 6: K-means de 3 cluster

La méthode k-means++ donne le résultat suivant :

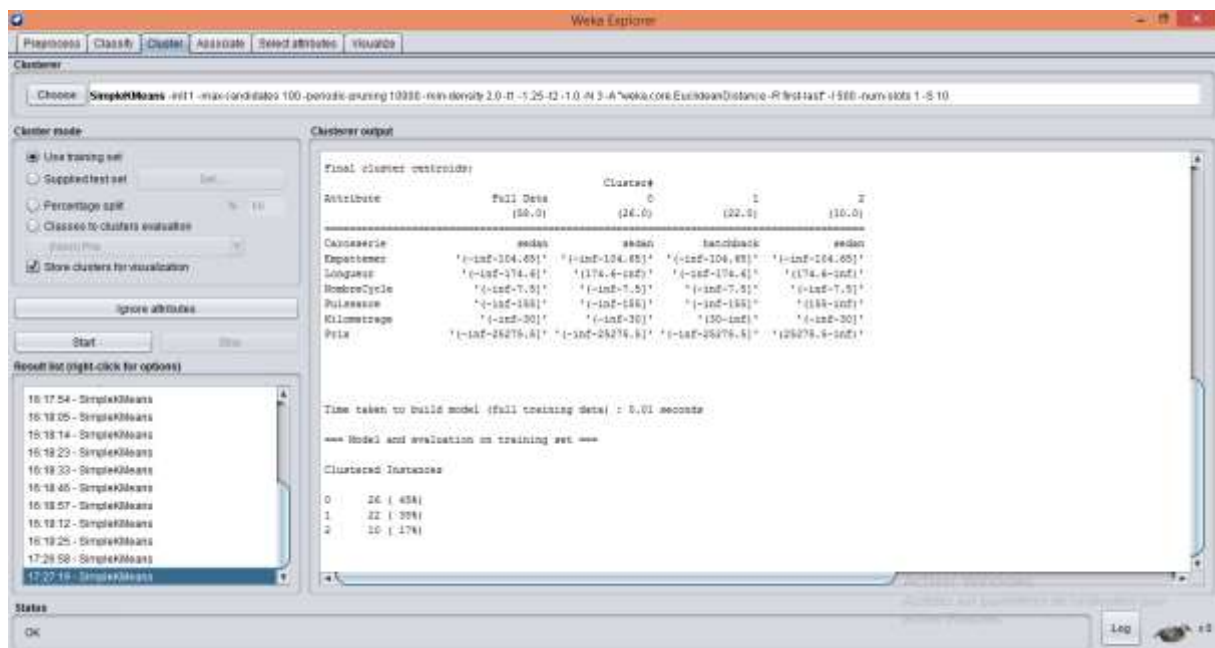


Figure 7: K-means++ de 3 cluster

On constate que La méthode k-means ++ est meilleur que k-means car le nombre des éléments dans le bloc .

Question 7 :

En fait le clustering avec les différents distances CAH .

En commence par la Distance la plus connu :

- distance euclidienne

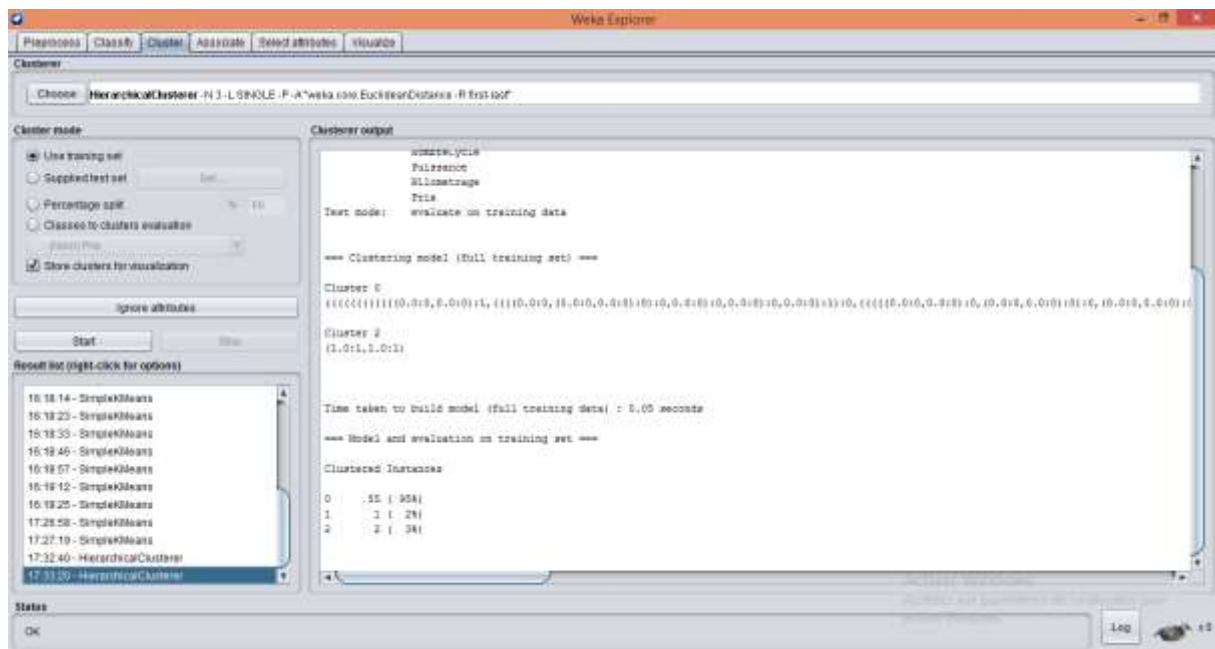


Figure 8: Distance euclidienne

La figure 8 montre que la résultat obtenu est très différent que le résultat obtenu par K-means.

- Distance Manhattan :

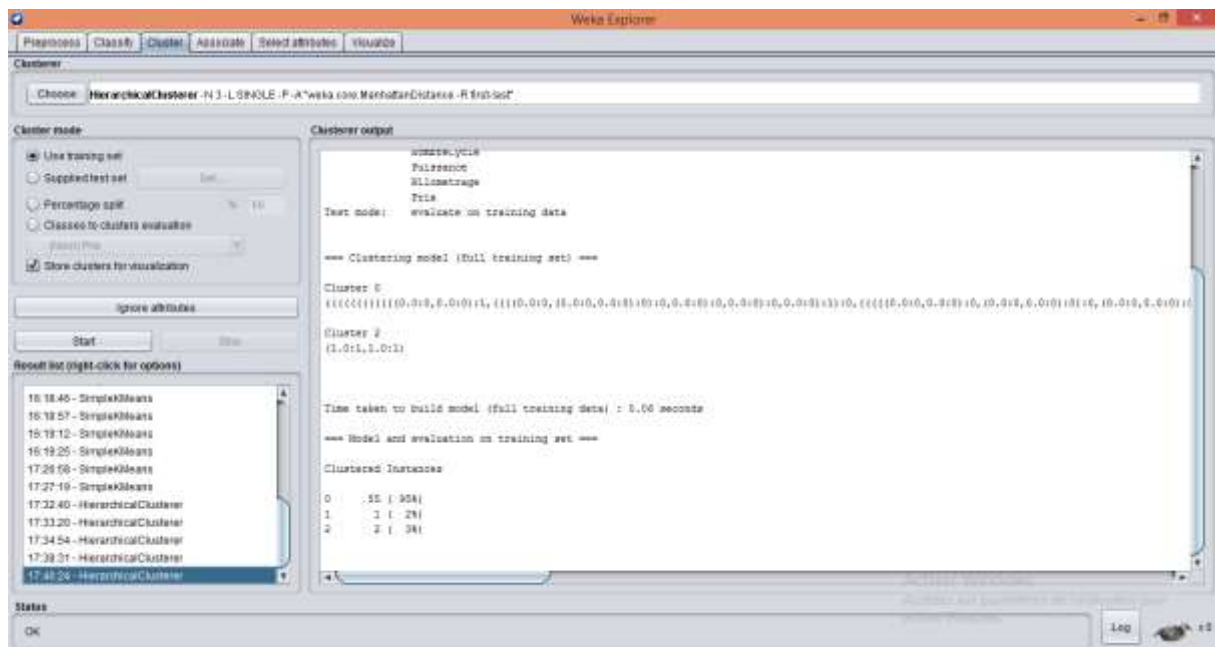


Figure 9: Distance Manhattan

Cette distance est similaire à la distance précédente donc elle est différente au K-means.

- [illegible]

Même concept que la distance précédente.

The screenshot shows the Weka Explorer interface with the following details:

- Buttons at the top:** Preprocess, Classify, Cluster, Associate, Select attributes, Visualize.
- Clusterer dropdown:** Set to "HierarchicalClusterer".
- Cluster mode section:**
 - ☒ Use training set
 - ☐ Supply test set
 - ☐ Percentage split
 - ☐ Classes to classify evaluation
 - ☒ Show clusters for visualization
 -
 -
- Result list (right-click for options):**
 - 16.18.23 - SimpleKMeans
 - 16.18.33 - SimpleKMeans
 - 16.18.46 - SimpleKMeans
 - 16.19.57 - SimpleKMeans
 - 16.19.12 - SimpleKMeans
 - 16.19.25 - SimpleKMeans
 - 17.25.58 - SimpleKMeans
 - 17.27.10 - SimpleKMeans
 - 17.32.40 - HierarchicalClusterer
 - 17.33.20 - HierarchicalClusterer
 - 17.34.54 - HierarchicalClusterer
- Cluster output window:**

```

Test mode:      J1C1
               evaluate on training data

== Clustering model (full training set) ==

Cluster 0
(0.0:0.0,0.0:0)

Cluster 1
(0.0:0.0,0.0:0)(1.0,0.0:0)

Cluster 2
((((((((((((((((((0.0:0.0,0.0:0):0,(0.0:0.0:0):0:(0.0,0.0:0):0),(0.0:0.0:0):0,0.0:0):1,(1.0:0,0.0:0):0):0,0.0:0)))

Time taken to build model (full training set) : 0.02 seconds

== Model and evaluation on training set ==

Clustered Instances
0       2 ( 3%)
1       8 ( 9%)
2     53 ( 91%)
    
```
- Status bar:** OK

Figure 11: distance chebyshev

- Distance filtered distance :

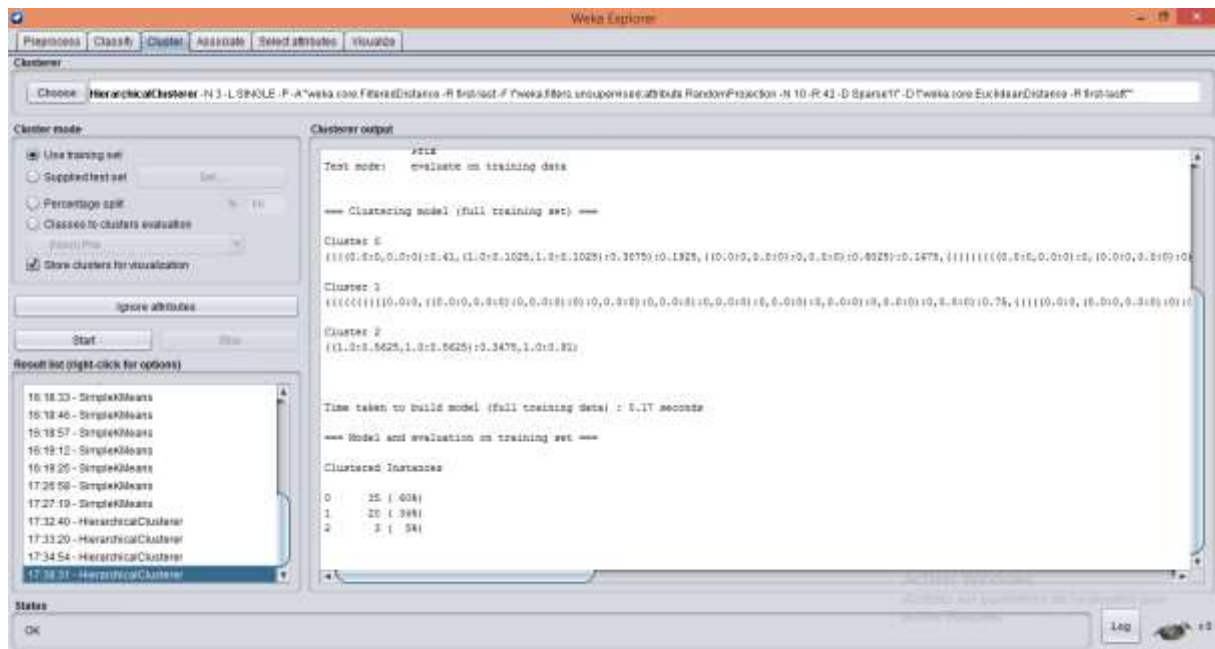


Figure 12: filtered distance

Cette distance est le meilleur car il est les plus proche au résultat de K-means.

Question 8 :

Pour améliorer les résultats de clustering je propose deux solutions :

- Augmenter les données d'entrées (fichier ARFF)
- Diminuer le nombre des attributs utilisés dans ce fichier.

Partie 2 : classification

In [1]:

```
import pandas as pd
import matplotlib.pyplot as plt
```

In []:

Je commence par l'importation de deux bibliothèques pandas et matplotlib.

In [2]:

```
file=pd.read_excel('fichiermodifier.xlsx')
```

Ensuite, je dois transformer les deux colonnes Carrosserie et Nombre-cylindres du fichier CaractéristiqueVéhicules en chiffre : 1 pour Carrosserie de type convertible 2 pour Carrosserie de type hatchback 3 pour Carrosserie de type sedan 4 pour Carrosserie de type wagon 5 pour Carrosserie de type hardtop

et après et après l'importation du ce dernier dans le Jupiter.

In [3]:

```
file.head()
```

Out[3]:

	Identifiant	Carrosserie	Empattement	longueur	Nombre-cylindres	Puissance	KilometrageMoyen	Prix
0	0	1	88.6	168.8	4	111	21	13495
1	1	1	88.6	168.8	4	111	21	16500
2	2	2	94.5	171.2	6	154	19	16500
3	3	3	99.8	176.6	4	102	24	13950
4	4	3	99.4	176.6	5	115	18	17450

Maintenant, je choisi le deux variables X qui correspond au caractéristiques et la cible y qui correspond a la prix.

In [6]:

```
X = file.iloc[:,1:7]
y = file.iloc[:,-1]
```

In [7]:

```
X
```

Out[7]:

	Carrosserie	Empattement	longueur	Nombre-cylindres	Puissance	KilometrageMoyen
0	1	88.6	168.8	4	111	21
1	1	88.6	168.8	4	111	21
2	2	94.5	171.2	6	154	19
3	3	99.8	176.6	4	102	24
4	3	99.4	176.6	5	115	18
5	3	99.8	177.3	5	110	19

6	Carrosserie	Empattement	Longueur	Nombre-cylindres	Puissance	KilometrageMoyen
7	3	101.2	176.8	4	101	23
8	3	101.2	176.8	4	101	23
9	3	101.2	176.8	6	121	21
10	3	103.5	189.0	6	182	16
11	3	103.5	193.8	6	182	16
12	3	110.0	197.0	6	182	15
13	2	88.4	141.1	3	48	47
14	2	94.5	155.9	4	70	38
15	3	94.5	158.8	4	70	38
16	2	93.7	157.3	4	68	31
17	2	93.7	157.3	4	68	31
18	4	96.5	157.1	4	76	30
19	3	96.5	175.4	4	101	24
20	3	96.5	169.1	4	100	25
21	3	94.3	170.7	4	78	24
22	3	113.0	199.6	6	176	15
23	3	113.0	199.6	6	176	15
24	3	102.0	191.7	13	262	13
25	2	93.1	159.1	4	68	30
26	2	93.1	159.1	4	68	31
27	2	93.1	159.1	4	68	31
28	2	95.3	169.0	2	101	17
29	3	104.9	175.0	4	72	31
30	3	110.0	190.9	5	123	22
31	4	110.0	190.9	5	123	22
32	3	120.9	208.1	8	184	14
33	5	112.0	199.2	8	184	14
34	2	93.7	157.3	4	68	37
35	2	93.7	157.3	4	68	31
36	3	96.3	172.4	4	88	25
37	3	96.3	172.4	4	88	25
38	3	94.5	165.3	4	55	45
39	3	94.5	165.3	4	69	31
40	3	94.5	165.3	4	69	31
41	2	94.5	170.2	4	69	31
42	3	100.4	184.6	6	152	19
43	5	89.5	168.9	6	207	17
44	1	89.5	168.9	6	207	17
45	2	95.7	158.7	4	62	35
46	2	95.7	158.7	4	62	31
47	2	95.7	158.7	4	62	31
48	4	95.7	169.7	4	62	31
49	4	95.7	169.7	4	62	27
50	4	95.7	169.7	4	62	27

51	Carrosserie	Empattement	longueur	Nombre-cylindres	Puissance	KilometrageMoyen
52	3	97.3	171.7	4	52	37
53	3	97.3	171.7	4	85	27
54	3	97.3	171.7	4	52	37
55	3	97.3	171.7	4	100	26
56	3	104.3	188.8	4	114	23
57	4	104.3	188.8	4	114	23

In [8]:

y

Out[8]:

- 013495
- 116500
- 216500
- 313950
- 417450
- 515250
- 618920
- 716430
- 816925
- 920970
- 1030760
- 1141315
- 1236880
- 135151
- 146295
- 156575
- 166377
- 176229
- 187295
- 1912945
- 2010345
- 216785
- 2232250
- 2335550
- 2436000
- 255195
- 266095
- 276795
- 2811845
- 2918344
- 3025552
- 3128248
- 3240960
- 3345400
- 345389
- 356189
- 366989
- 378189
- 387099
- 396649
- 406849
- 417349
- 4213499
- 4334028
- 4437028
- 455348
- 466338
- 476488
- 486918
- 497898
- 508778
- 5115750
- 527775
- 537975

```
54      7995
55      9995
56     12940
57     13415
Name: Prix, dtype: int64
```

Comme montre les teste ci-dessus que l'importation du fichier et la création de deux variables X et y a été bien passée en commence le test des algorithmes.

SVM :

In [9]:

```
from sklearn import svm
from sklearn.metrics import plot_confusion_matrix
from sklearn.model_selection import train_test_split
X_train1, X_test1, y_train1, y_test1 = train_test_split(X, y, test_size=0.5, random_state=0)
v_clf = svm.SVC()
v_clf.fit(X_train1, y_train1)
```

Out[9]:

SVC()

In [10]:

```
v_clf.predict(X_test1)
```

Out[10]:

```
array([ 6377,  6377, 15750,  5195, 36880, 15750, 36880,  6488, 35550,
        8778, 20970, 36880,  6649,  6649, 35550,  6649,  6989, 16925,
        6377, 13415, 10345, 13415,  7099,  6377,  7099,  5195,  6377,
       13950, 13415], dtype=int64)
```

In [11]:

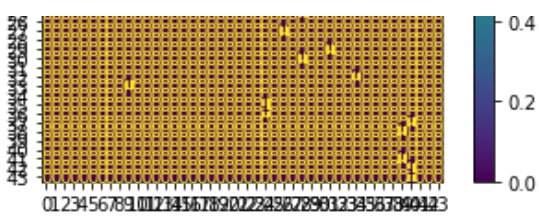
```
#Pour voit la différence entre les algorithmes je dois effectuer à chaque fois la matrice
de confusion et le score du chacun.
#Matrice de confusion de l'algorithmme SVM
plot_confusion_matrix(v_clf, X_test1, y_test1)
plt.show()
```

```
-----
ValueError                                Traceback (most recent call last)
<ipython-input-11-b92ccdae29fd> in <module>
      1 #Pour voit la différence entre les algorithmes je dois effectuer à chaque fois la
matrice de confusion et le score du chacun.
----> 2 plot_confusion_matrix(v_clf, X_test1, y_test1)
      3 plt.show()
```

```
C:\ProgramData\Anaconda3\lib\site-packages\sklearn\utils\validation.py in inner_f(*args,
**kwargs)
     70         FutureWarning)
     71     kwargs.update({k: arg for k, arg in zip(sig.parameters, args)})
----> 72     return f(**kwargs)
     73     return inner_f
     74
```

```
C:\ProgramData\Anaconda3\lib\site-packages\sklearn\metrics\_plot\confusion_matrix.py in p
lot_confusion_matrix(estimator, X, y_true, labels, sample_weight, normalize, display_labe
ls, include_values, xticks_rotation, values_format, cmap, ax)
    229     disp = ConfusionMatrixDisplay(confusion_matrix=cm,
    230                                     display_labels=display_labels)
--> 231     return disp.plot(include_values=include_values,
    232                       cmap=cmap, ax=ax, xticks_rotation=xticks_rotation,
    233                       values_format=values_format)
```

```
C:\ProgramData\Anaconda3\lib\site-packages\sklearn\utils\validation.py in inner_f(*args,
```

On peut constater que plus la couleur tend vers la couleur jaune l'algorithme est plus performant.

In [13]:

```
v_clf.score(X_train1,y_train1)
#score de l'algorithme SVM
```

Out[13]:

```
0.8620689655172413
```

LogisticRegression :

In [14]:

```
from sklearn.linear_model import LogisticRegression
from sklearn.model_selection import train_test_split
```

```
X_train2, X_test2, y_train2, y_test2 = train_test_split(X, y, test_size=0.5, random_state=0)
lreg = LogisticRegression().fit(X_train2, y_train2)
```

C:\ProgramData\Anaconda3\lib\site-packages\sklearn\linear_model_logistic.py:762: ConvergenceWarning: lbfgs failed to converge (status=1):
STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.

Increase the number of iterations (max_iter) or scale the data as shown in:

<https://scikit-learn.org/stable/modules/preprocessing.html>

Please also refer to the documentation for alternative solver options:

https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression

```
n_iter_i = _check_optimize_result(
```

In [15]:

```
lreg.predict(X_test2)
```

Out[15]:

```
array([ 6377,  6229, 15750,  5195, 36880, 15750, 36880,  6488, 35550,
        8778, 20970, 36880,  6338, 6649, 35550,  6649,  6989, 16925,
        6377, 20970, 18920, 13415,  7099,  6377,  7099,  5195,  6229,
       13415, 20970], dtype=int64)
```

In [16]:

```
plot_confusion_matrix(lreg, X_test2, y_test2)
plt.show()
```

ValueError Traceback (most recent call last)

<ipython-input-16-30e818939c7a> in <module>

```
----> 1 plot_confusion_matrix(lreg, X_test2, y_test2)
      2 plt.show()
```

C:\ProgramData\Anaconda3\lib\site-packages\sklearn\utils\validation.py in inner_f(*args, **kwargs)

```
70                                     FutureWarning)
```

```
71     kwargs.update({k: arg for k, arg in zip(sig.parameters, args)})
```

```
----> 72     return f(**kwargs)
```

```
73     return inner_f
```

```
74
```

C:\ProgramData\Anaconda3\lib\site-packages\sklearn\metrics_plot_confusion_matrix.py in

```

C:\ProgramData\Anaconda3\lib\site-packages\sklearn\metrics\_plot\confusion_matrix.py in p
lot_confusion_matrix(estimator, X, y_true, labels, sample_weight, normalize, display_labe
ls, include_values, xticks_rotation, values_format, cmap, ax)
    229     disp = ConfusionMatrixDisplay(confusion_matrix=cm,
    230                                     display_labels=display_labels)
--> 231     return disp.plot(include_values=include_values,
    232                       cmap=cmap, ax=ax, xticks_rotation=xticks_rotation,
    233                       values_format=values_format)

```

```

C:\ProgramData\Anaconda3\lib\site-packages\sklearn\utils\validation.py in inner_f(*args,
**kwargs)
    70         FutureWarning)
    71     kwargs.update({k: arg for k, arg in zip(sig.parameters, args)})
---> 72     return f(**kwargs)
    73     return inner_f
    74

```

```

C:\ProgramData\Anaconda3\lib\site-packages\sklearn\metrics\_plot\confusion_matrix.py in p
lot(self, include_values, cmap, xticks_rotation, values_format, ax)
    118
    119     fig.colorbar(self.im_, ax=ax)
--> 120     ax.set(xticks=np.arange(n_classes),
    121           yticks=np.arange(n_classes),
    122           xticklabels=display_labels,

```

```

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\artist.py in set(self, **kwargs)
   1111     if move_color_to_start:
   1112         kwargs = {"color": kwargs.pop("color"), **kwargs}
-> 1113     return self.update(kwargs)
   1114
   1115     def findobj(self, match=None, include_self=True):

```

```

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\artist.py in update(self, props)
    996         raise AttributeError(f"{type(self).__name__!r} object "
    997                               f"has no property {k!r}")
--> 998         ret.append(func(v))
    999     if ret:
   1000         self.pchanged()

```

```

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\axes\_base.py in wrapper(self, *arg
s, **kwargs)
    61
    62     def wrapper(self, *args, **kwargs):
---> 63         return get_method(self)(*args, **kwargs)
    64
    65     wrapper.__module__ = owner.__module__

```

```

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\cbook\deprecation.py in wrapper(*ar
gs, **kwargs)
    449         "parameter will become keyword-only %(removal)s.",
    450         name=name, obj_type=f"parameter of {func.__name__}()")
--> 451     return func(*args, **kwargs)
    452
    453     return wrapper

```

```

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\axis.py in _set_ticklabels(self, la
bels, fontdict, minor, **kwargs)
   1791     if fontdict is not None:
   1792         kwargs.update(fontdict)
-> 1793     return self.set_ticklabels(labels, minor=minor, **kwargs)
   1794
   1795     @cbook._make_keyword_only("3.2", "minor")

```

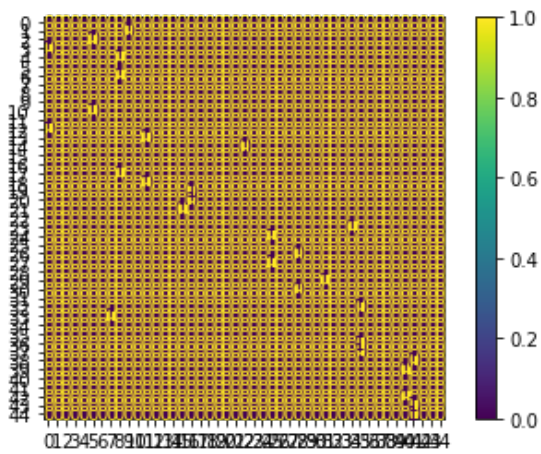
```

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\axis.py in set_ticklabels(self, tic
klables, minor, **kwargs)
   1712         # remove all tick labels, so only error for > 0 ticklabels
   1713         if len(locator.locs) != len(ticklabels) and len(ticklabels) != 0:
-> 1714             raise ValueError(
   1715                 "The number of FixedLocator locations"
   1716                 f" ({len(locator.locs)}, usually from a call to"

```

ValueError: The number of FixedLocator locations (45), usually from a call to set_ticks, does not match the number of ticklabels (20)

does not match the number of ticklabels (29).



In [17]:

```
lreg.score(X_train2,y_train2)
```

Out[17]:

0.8620689655172413

Naïve bayes :

In [18]:

```
from sklearn.naive_bayes import GaussianNB
from sklearn.model_selection import train_test_split

X_train3, X_test3, y_train3, y_test3 = train_test_split(X, y, test_size=0.5, random_state=0)
nvb = GaussianNB().fit(X_train3,y_train3)
```

In [19]:

```
nvb.score(X_train3,y_train3)
```

Out[19]:

0.8620689655172413

In [20]:

```
nvb.predict(X_test3)
```

Out[20]:

```
array([ 6229,  6229, 15750,  5195, 36880, 15750, 36880,  6338, 35550,
        7898, 20970, 36880,  6649,  6649, 35550,  6649,  6989, 16925,
        6229, 13415, 10345, 13415,  7099,  6229,  7099,  5195,  6229,
       13950, 13415], dtype=int64)
```

In [21]:

```
plot_confusion_matrix(nvb, X_test3, y_test3)
plt.show()
```

ValueError Traceback (most recent call last)

```
<ipython-input-21-1bfe02f43cc6> in <module>
----> 1 plot_confusion_matrix(nvb, X_test3, y_test3)
      2 plt.show()
```

```
C:\ProgramData\Anaconda3\lib\site-packages\sklearn\utils\validation.py in inner_f(*args,
**kwargs)
```

```
70         FutureWarning)
71         kwargs.update({k: arg for k, arg in zip(sig.parameters, args)})
---> 72         return f(**kwargs)
```

```

73     return inner_f
74

C:\ProgramData\Anaconda3\lib\site-packages\sklearn\metrics\_plot\confusion_matrix.py in plot_confusion_matrix(estimator, X, y_true, labels, sample_weight, normalize, display_labels, include_values, xticks_rotation, values_format, cmap, ax)
229     disp = ConfusionMatrixDisplay(confusion_matrix=cm,
230                                   display_labels=display_labels)
--> 231     return disp.plot(include_values=include_values,
232                     cmap=cmap, ax=ax, xticks_rotation=xticks_rotation,
233                     values_format=values_format)

C:\ProgramData\Anaconda3\lib\site-packages\sklearn\utils\validation.py in inner_f(*args, **kwargs)
70         FutureWarning)
71     kwargs.update({k: arg for k, arg in zip(sig.parameters, args)})
---> 72     return f(**kwargs)
73     return inner_f
74

C:\ProgramData\Anaconda3\lib\site-packages\sklearn\metrics\_plot\confusion_matrix.py in plot(self, include_values, cmap, xticks_rotation, values_format, ax)
118
119     fig.colorbar(self.im_, ax=ax)
--> 120     ax.set(xticks=np.arange(n_classes),
121           yticks=np.arange(n_classes),
122           xticklabels=display_labels,

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\artist.py in set(self, **kwargs)
1111     if move_color_to_start:
1112         kwargs = {"color": kwargs.pop("color"), **kwargs}
-> 1113     return self.update(kwargs)
1114
1115     def findobj(self, match=None, include_self=True):

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\artist.py in update(self, props)
996         raise AttributeError(f"{type(self).__name__!r} object "
997                               f"has no property {k!r}")
--> 998         ret.append(func(v))
999     if ret:
1000         self.pchanged()

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\axes\_base.py in wrapper(self, *args, **kwargs)
61
62     def wrapper(self, *args, **kwargs):
---> 63         return get_method(self)(*args, **kwargs)
64
65     wrapper.__module__ = owner.__module__

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\cbook\deprecation.py in wrapper(*args, **kwargs)
449         "parameter will become keyword-only %(removal)s.",
450         name=name, obj_type=f"parameter of {func.__name__}()")
--> 451     return func(*args, **kwargs)
452
453     return wrapper

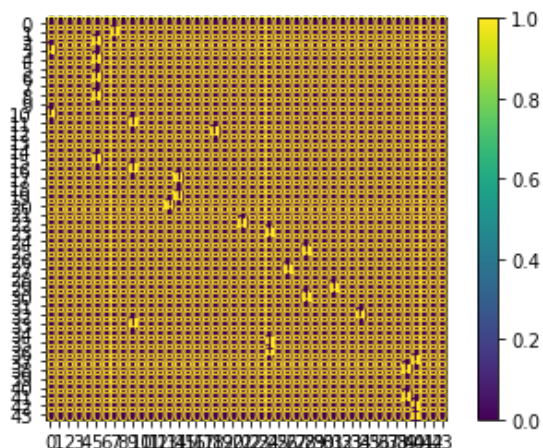
C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\axis.py in _set_ticklabels(self, labels, fontdict, minor, **kwargs)
1791     if fontdict is not None:
1792         kwargs.update(fontdict)
-> 1793     return self.set_ticklabels(labels, minor=minor, **kwargs)
1794
1795     @cbook._make_keyword_only("3.2", "minor")

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\axis.py in set_ticklabels(self, ticklabels, minor, **kwargs)
1712     # remove all tick labels, so only error for > 0 ticklabels
1713     if len(locator.locs) != len(ticklabels) and len(ticklabels) != 0:
-> 1714         raise ValueError(
1715             "The number of FixedLocator locations"
```

1716

f" ({len(locator.locs)}), usually from a call to"

ValueError: The number of FixedLocator locations (44), usually from a call to `set_ticks`, does not match the number of ticklabels (29).



In [23]:

```
file.iloc[44,:]
```

Out[23]:

```
Identifiant      62.0
Carrosserie      1.0
Empattement     89.5
longueur       168.9
Nombre-cylindres  6.0
Puissance       207.0
KilometrageMoyen 17.0
Prix           37028.0
Name: 44, dtype: float64
```

KNN :

In [24]:

```
from sklearn.neighbors import KNeighborsClassifier
```

```
X_train3, X_test3, y_train3, y_test3 = train_test_split(X, y, test_size=0.5, random_state=0)
knn = KNeighborsClassifier(n_neighbors = 1).fit(X_train3,y_train3)
```

In [25]:

```
knn.score(X_train3,y_train3)
```

Out[25]:

0.8620689655172413

In [26]:

```
knn.predict(X_test3)
```

Out[26]:

```
array([ 6377,  6377, 15750,  5195, 36880, 15750, 36880,  6338, 35550,
        7898, 20970, 36880,  6649,  6649, 35550,  6649,  6989, 16925,
        6377, 13415, 10345, 13415,  7099,  6377,  7099,  5195,  6377,
       13950, 13415], dtype=int64)
```

In [27]:

```
plot_confusion_matrix(knn, X_test3, y_test3)
plt.show()
```

```

-----
ValueError                                Traceback (most recent call last)
<ipython-input-27-70dbb7b96b1a> in <module>
----> 1 plot_confusion_matrix(knn, X_test3, y_test3)
      2 plt.show()

C:\ProgramData\Anaconda3\lib\site-packages\sklearn\utils\validation.py in inner_f(*args,
**kwargs)
      70             FutureWarning)
      71         kwargs.update({k: arg for k, arg in zip(sig.parameters, args)})
----> 72         return f(**kwargs)
      73     return inner_f
      74

C:\ProgramData\Anaconda3\lib\site-packages\sklearn\metrics\_plot\confusion_matrix.py in p
lot_confusion_matrix(estimator, X, y_true, labels, sample_weight, normalize, display_labe
ls, include_values, xticks_rotation, values_format, cmap, ax)
      229     disp = ConfusionMatrixDisplay(confusion_matrix=cm,
      230                                   display_labels=display_labels)
--> 231     return disp.plot(include_values=include_values,
      232                     cmap=cmap, ax=ax, xticks_rotation=xticks_rotation,
      233                     values_format=values_format)

C:\ProgramData\Anaconda3\lib\site-packages\sklearn\utils\validation.py in inner_f(*args,
**kwargs)
      70             FutureWarning)
      71         kwargs.update({k: arg for k, arg in zip(sig.parameters, args)})
----> 72         return f(**kwargs)
      73     return inner_f
      74

C:\ProgramData\Anaconda3\lib\site-packages\sklearn\metrics\_plot\confusion_matrix.py in p
lot(self, include_values, cmap, xticks_rotation, values_format, ax)
      118
      119     fig.colorbar(self.im_, ax=ax)
--> 120     ax.set(xticks=np.arange(n_classes),
      121           yticks=np.arange(n_classes),
      122           xticklabels=display_labels,

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\artist.py in set(self, **kwargs)
     1111     if move_color_to_start:
     1112         kwargs = {"color": kwargs.pop("color"), **kwargs}
-> 1113     return self.update(kwargs)
     1114
     1115     def findobj(self, match=None, include_self=True):

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\artist.py in update(self, props)
      996         raise AttributeError(f"{type(self).__name__!r} object "
      997                               f"has no property {k!r}")
--> 998         ret.append(func(v))
      999     if ret:
     1000         self.pchanged()

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\axes\_base.py in wrapper(self, *arg
s, **kwargs)
      61
      62     def wrapper(self, *args, **kwargs):
--> 63         return get_method(self)(*args, **kwargs)
      64
      65     wrapper.__module__ = owner.__module__

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\cbook\deprecation.py in wrapper(*ar
gs, **kwargs)
      449         "parameter will become keyword-only %(removal)s.",
      450         name=name, obj_type=f"parameter of {func.__name__}()")
--> 451     return func(*args, **kwargs)
      452
      453     return wrapper

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\axis.py in _set_ticklabels(self, la
bels, fontdict, minor, **kwargs)
     1791     if fontdict is not None:

```

```
ValueError                                Traceback (most recent call last)
<ipython-input-31-93a975c7560e> in <module>
----> 1 plot_confusion_matrix(ard, X_test4, y_test4)
      2 plt.show()
```

```

C:\ProgramData\Anaconda3\lib\site-packages\sklearn\utils\validation.py in inner_f(*args,
**kwargs)
    70         FutureWarning)
    71         kwargs.update({k: arg for k, arg in zip(sig.parameters, args)})
--> 72         return f(**kwargs)
    73     return inner_f
    74

```

```

C:\ProgramData\Anaconda3\lib\site-packages\sklearn\metrics\_plot\confusion_matrix.py in p
lot_confusion_matrix(estimator, X, y_true, labels, sample_weight, normalize, display_labe
ls, include_values, xticks_rotation, values_format, cmap, ax)
    229     disp = ConfusionMatrixDisplay(confusion_matrix=cm,
    230                                     display_labels=display_labels)
--> 231     return disp.plot(include_values=include_values,
    232                     cmap=cmap, ax=ax, xticks_rotation=xticks_rotation,
    233                     values_format=values_format)

```

```

C:\ProgramData\Anaconda3\lib\site-packages\sklearn\utils\validation.py in inner_f(*args,
**kwargs)
    70         FutureWarning)
    71         kwargs.update({k: arg for k, arg in zip(sig.parameters, args)})
--> 72         return f(**kwargs)
    73     return inner_f
    74

```

```

C:\ProgramData\Anaconda3\lib\site-packages\sklearn\metrics\_plot\confusion_matrix.py in p
lot(self, include_values, cmap, xticks_rotation, values_format, ax)
    118
    119     fig.colorbar(self.im_, ax=ax)
--> 120     ax.set(xticks=np.arange(n_classes),
    121           yticks=np.arange(n_classes),
    122           xticklabels=display_labels,

```

```

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\artist.py in set(self, **kwargs)
    1111     if move_color_to_start:
    1112         kwargs = {"color": kwargs.pop("color"), **kwargs}
-> 1113     return self.update(kwargs)
    1114
    1115     def findobj(self, match=None, include_self=True):

```

```

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\artist.py in update(self, props)
    996         raise AttributeError(f"{type(self).__name__!r} object "
    997                               f"has no property {k!r}")
--> 998         ret.append(func(v))
    999     if ret:
    1000         self.pchanged()

```

```

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\axes\_base.py in wrapper(self, *arg
s, **kwargs)
    61
    62     def wrapper(self, *args, **kwargs):
--> 63         return get_method(self)(*args, **kwargs)
    64
    65     wrapper.__module__ = owner.__module__

```

```

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\cbook\deprecation.py in wrapper(*ar
gs, **kwargs)
    449         "parameter will become keyword-only %(removal)s.",
    450         name=name, obj_type=f"parameter of {func.__name__}()")
--> 451     return func(*args, **kwargs)
    452
    453     return wrapper

```

```

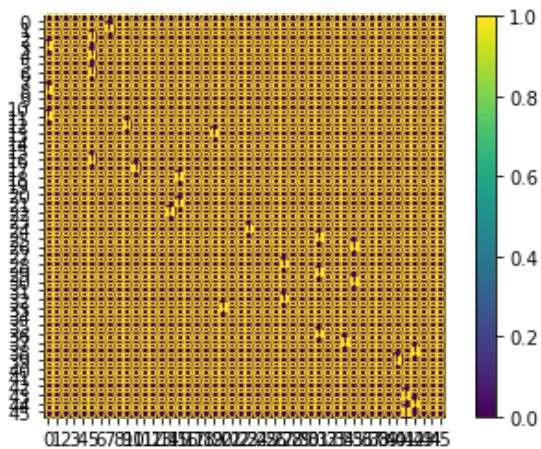
C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\axis.py in _set_ticklabels(self, la
bels, fontdict, minor, **kwargs)
    1791     if fontdict is not None:
    1792         kwargs.update(fontdict)
-> 1793     return self.set_ticklabels(labels, minor=minor, **kwargs)
    1794
    1795     @cbook._make_keyword_only("3.2", "minor")

```



```
C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\axis.py in set_ticklabels(self, ticklabels, minor, **kwargs)
    1712         # remove all tick labels, so only error for > 0 ticklabels
    1713         if len(locator.locs) != len(ticklabels) and len(ticklabels) != 0:
-> 1714             raise ValueError(
    1715                 "The number of FixedLocator locations"
    1716                 f" ({len(locator.locs)}), usually from a call to"
```

ValueError: The number of FixedLocator locations (46), usually from a call to set_ticks, does not match the number of ticklabels (29).



Multilayer Perceptron :

In [32]:

```
from sklearn.neural_network import MLPClassifier
X_train5, X_test5, y_train5, y_test5 = train_test_split(X, y, test_size=0.5, random_state=0)
mlp = MLPClassifier(solver='lbfgs', alpha=1e-5, hidden_layer_sizes=(5, 2), random_state=1).fit(X_train5, y_train5)
```

In [33]:

```
mlp.score(X_train5, y_train5)
```

Out[33]:

```
0.06896551724137931
```

In [34]:

```
mlp.predict(X_test5)
```

Out[34]:

```
array([[34028, 7099, 6989, 34028, 6989, 6989, 6989, 7099, 6989,
        34028, 6989, 6989, 34028, 34028, 6989, 34028, 6989, 6989,
        7099, 6989, 6989, 6989, 34028, 34028, 34028, 34028, 7099,
        6989, 6989], dtype=int64)
```

In [35]:

```
plot_confusion_matrix(mlp, X_test5, y_test5)
plt.show()
```

```
-----
ValueError                                Traceback (most recent call last)
<ipython-input-35-1b96033ac089> in <module>
----> 1 plot_confusion_matrix(mlp, X_test5, y_test5)
      2 plt.show()
```

```
C:\ProgramData\Anaconda3\lib\site-packages\sklearn\utils\validation.py in inner_f(*args, **kwargs)
    70         FutureWarning)
    71         kwargs.update({k: arg for k, arg in zip(sig.parameters, args)})
```



```

---> 72         return f(**kwargs)
      73     return inner_f
      74

```

C:\ProgramData\Anaconda3\lib\site-packages\sklearn\metrics_plot\confusion_matrix.py in plot_confusion_matrix(estimator, X, y_true, labels, sample_weight, normalize, display_labels, include_values, xticks_rotation, values_format, cmap, ax)

```

      229     disp = ConfusionMatrixDisplay(confusion_matrix=cm,
      230                                     display_labels=display_labels)
--> 231     return disp.plot(include_values=include_values,
      232                     cmap=cmap, ax=ax, xticks_rotation=xticks_rotation,
      233                     values_format=values_format)

```

C:\ProgramData\Anaconda3\lib\site-packages\sklearn\utils\validation.py in inner_f(*args, **kwargs)

```

      70         FutureWarning)
      71     kwargs.update({k: arg for k, arg in zip(sig.parameters, args)})
---> 72     return f(**kwargs)
      73     return inner_f
      74

```

C:\ProgramData\Anaconda3\lib\site-packages\sklearn\metrics_plot\confusion_matrix.py in plot(self, include_values, cmap, xticks_rotation, values_format, ax)

```

      118
      119     fig.colorbar(self.im_, ax=ax)
--> 120     ax.set(xticks=np.arange(n_classes),
      121           yticks=np.arange(n_classes),
      122           xticklabels=display_labels,

```

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\artist.py in set(self, **kwargs)

```

     1111     if move_color_to_start:
     1112         kwargs = {"color": kwargs.pop("color"), **kwargs}
-> 1113     return self.update(kwargs)
     1114
     1115     def findobj(self, match=None, include_self=True):

```

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\artist.py in update(self, props)

```

      996         raise AttributeError(f"{type(self).__name__!r} object "
      997                               f"has no property {k!r}")
--> 998         ret.append(func(v))
      999     if ret:
     1000         self.pchanged()

```

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\axes_base.py in wrapper(self, *args, **kwargs)

```

      61
      62     def wrapper(self, *args, **kwargs):
---> 63         return get_method(self)(*args, **kwargs)
      64
      65     wrapper.__module__ = owner.__module__

```

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\cbook\deprecation.py in wrapper(*args, **kwargs)

```

      449         "parameter will become keyword-only %(removal)s.",
      450         name=name, obj_type=f"parameter of {func.__name__}()")
--> 451     return func(*args, **kwargs)
      452
      453     return wrapper

```

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\axis.py in _set_ticklabels(self, labels, fontdict, minor, **kwargs)

```

     1791     if fontdict is not None:
     1792         kwargs.update(fontdict)
-> 1793     return self.set_ticklabels(labels, minor=minor, **kwargs)
     1794
     1795     @cbook._make_keyword_only("3.2", "minor")

```

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\axis.py in set_ticklabels(self, ticklabels, minor, **kwargs)

```

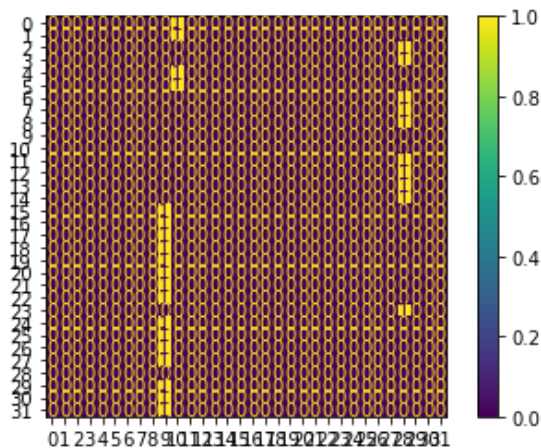
     1712         # remove all tick labels, so only error for > 0 ticklabels
     1713         if len(locator.locs) != len(ticklabels) and len(ticklabels) != 0:
-> 1714             raise ValueError(

```

```
1715
1716
```

```
"The number of FixedLocator locations"
f" ({len(locator.locs)}), usually from a call to"
```

ValueError: The number of FixedLocator locations (32), usually from a call to `set_ticks`, does not match the number of ticklabels (29).



Adaboost :

In [36]:

```
from sklearn.ensemble import AdaBoostClassifier
from sklearn.datasets import make_classification
X_train6, X_test6, y_train6, y_test6 = train_test_split(X, y, test_size=0.5, random_state=0)
ads = AdaBoostClassifier(n_estimators=100, random_state=0).fit(X_train6, y_train6)
```

In [37]:

```
ads.score(X_train6, y_train6)
```

Out[37]:

```
0.7241379310344828
```

In [38]:

```
ads.predict(X_test6)
```

Out[38]:

```
array([ 6229,  6229, 20970,  5195, 15750, 20970, 35550,  6338, 35550,
        7898, 16925, 15750,  6989,  6649, 35550, 12945,  6989, 16925,
        6229, 13415, 16925, 13415,  7898,  6649,  7898,  5195,  6649,
        16925, 13415], dtype=int64)
```

In [39]:

```
plot_confusion_matrix(ads, X_test6, y_test6)
plt.show()
```

ValueError Traceback (most recent call last)

```
<ipython-input-39-bd23ea934147> in <module>
----> 1 plot_confusion_matrix(ads, X_test6, y_test6)
      2 plt.show()
```

C:\ProgramData\Anaconda3\lib\site-packages\sklearn\utils\validation.py in `inner_f(*args, **kwargs)`

```
70                                     FutureWarning)
71     kwargs.update({k: arg for k, arg in zip(sig.parameters, args)})
---> 72     return f(**kwargs)
73     return inner_f
74
```

C:\ProgramData\Anaconda3\lib\site-packages\sklearn\metrics\ plot\confusion matrix.py in p

```

lot_confusion_matrix(estimator, X, y_true, labels, sample_weight, normalize, display_labels,
include_values, xticks_rotation, values_format, cmap, ax)
229     disp = ConfusionMatrixDisplay(confusion_matrix=cm,
230                                   display_labels=display_labels)
--> 231     return disp.plot(include_values=include_values,
232                      cmap=cmap, ax=ax, xticks_rotation=xticks_rotation,
233                      values_format=values_format)

```

```

C:\ProgramData\Anaconda3\lib\site-packages\sklearn\utils\validation.py in inner_f(*args,
**kwargs)
70         FutureWarning)
71     kwargs.update({k: arg for k, arg in zip(sig.parameters, args)})
---> 72     return f(**kwargs)
73     return inner_f
74

```

```

C:\ProgramData\Anaconda3\lib\site-packages\sklearn\metrics\_plot\confusion_matrix.py in plot(self, include_values, cmap, xticks_rotation, values_format, ax)
118
119     fig.colorbar(self.im_, ax=ax)
--> 120     ax.set(xticks=np.arange(n_classes),
121           yticks=np.arange(n_classes),
122           xticklabels=display_labels,

```

```

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\artist.py in set(self, **kwargs)
1111     if move_color_to_start:
1112         kwargs = {"color": kwargs.pop("color"), **kwargs}
-> 1113     return self.update(kwargs)
1114
1115     def findobj(self, match=None, include_self=True):

```

```

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\artist.py in update(self, props)
996         raise AttributeError(f"{type(self).__name__}!r object "
997                             f"has no property {k!r}")
--> 998         ret.append(func(v))
999     if ret:
1000         self.pchanged()

```

```

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\axes\_base.py in wrapper(self, *args,
**kwargs)
61
62     def wrapper(self, *args, **kwargs):
---> 63         return get_method(self)(*args, **kwargs)
64
65     wrapper.__module__ = owner.__module__

```

```

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\cbook\deprecation.py in wrapper(*args,
**kwargs)
449         "parameter will become keyword-only %(removal)s.",
450         name=name, obj_type=f"parameter of {func.__name__}()")
--> 451     return func(*args, **kwargs)
452
453     return wrapper

```

```

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\axis.py in _set_ticklabels(self, labels, fontdict, minor, **kwargs)
1791     if fontdict is not None:
1792         kwargs.update(fontdict)
-> 1793     return self.set_ticklabels(labels, minor=minor, **kwargs)
1794
1795     @cbook._make_keyword_only("3.2", "minor")

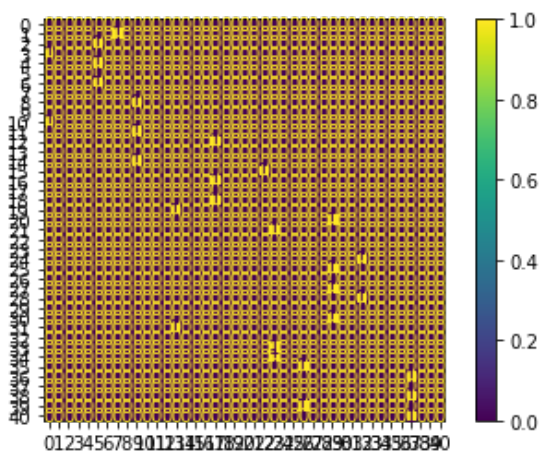
```

```

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\axis.py in set_ticklabels(self, ticklabels, minor, **kwargs)
1712         # remove all tick labels, so only error for > 0 ticklabels
1713         if len(locator.locs) != len(ticklabels) and len(ticklabels) != 0:
-> 1714             raise ValueError(
1715                 "The number of FixedLocator locations"
1716                 f" ({len(locator.locs)}), usually from a call to"

```

ValueError: The number of FixedLocator locations (41), usually from a call to set_ticks, does not match the number of ticklabels (29).



Random Forest :

In [40]:

```
from sklearn.ensemble import RandomForestClassifier
from sklearn.datasets import make_classification
X_train7, X_test7, y_train7, y_test7 = train_test_split(X, y, test_size=0.5, random_state=0)
rf = RandomForestClassifier(max_depth=2, random_state=0).fit(X_train7,y_train7)
```

In [41]:

```
rf.score(X_train7,y_train7)
```

Out[41]:

0.8275862068965517

In [43]:

```
rf.predict(X_test7)
```

Out[43]:

```
array([ 6229,  6229, 20970,  5195, 35550, 20970, 35550,  6488, 35550,
        8778, 12945, 20970,  6649,  6649, 35550,  6649,  6989, 16925,
        6229, 18920,  5151, 13415,  7099,  6377,  7099,  5195,  6649,
       12945, 36880], dtype=int64)
```

In [44]:

```
plot_confusion_matrix(rf, X_test7, y_test7)
plt.show()
```

ValueError Traceback (most recent call last)

```
<ipython-input-44-29fb098eae73> in <module>
----> 1 plot_confusion_matrix(rf, X_test7, y_test7)
      2 plt.show()
```

```
C:\ProgramData\Anaconda3\lib\site-packages\sklearn\utils\validation.py in inner_f(*args,
**kwargs)
    70         FutureWarning)
    71         kwargs.update({k: arg for k, arg in zip(sig.parameters, args)})
--> 72         return f(**kwargs)
    73     return inner_f
    74
```

```
C:\ProgramData\Anaconda3\lib\site-packages\sklearn\metrics\_plot\confusion_matrix.py in p
lot_confusion_matrix(estimator, X, y_true, labels, sample_weight, normalize, display_labe
ls, include_values, xticks_rotation, values_format, cmap, ax)
    229     disp = ConfusionMatrixDisplay(confusion_matrix=cm,
    230                                     display_labels=display_labels)
--> 231     return disp.plot(include_values=include_values,
```

```

332             cmap=cmap, ax=ax, xticks_rotation=xticks_rotation,
333             values_format=values_format)

C:\ProgramData\Anaconda3\lib\site-packages\sklearn\utils\validation.py in inner_f(*args,
**kwargs)
    70             FutureWarning)
    71             kwargs.update({k: arg for k, arg in zip(sig.parameters, args)})
--> 72             return f(**kwargs)
    73         return inner_f
    74

C:\ProgramData\Anaconda3\lib\site-packages\sklearn\metrics\_plot\confusion_matrix.py in p
lot(self, include_values, cmap, xticks_rotation, values_format, ax)
    118
    119         fig.colorbar(self.im_, ax=ax)
--> 120         ax.set(xticks=np.arange(n_classes),
    121               yticks=np.arange(n_classes),
    122               xticklabels=display_labels,

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\artist.py in set(self, **kwargs)
   1111         if move_color_to_start:
   1112             kwargs = {"color": kwargs.pop("color"), **kwargs}
-> 1113         return self.update(kwargs)
   1114
   1115         def findobj(self, match=None, include_self=True):

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\artist.py in update(self, props)
    996             raise AttributeError(f"{type(self).__name__!r} object "
    997                                   f"has no property {k!r}")
--> 998             ret.append(func(v))
    999         if ret:
   1000             self.pchanged()

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\axes\_base.py in wrapper(self, *arg
s, **kwargs)
    61
    62         def wrapper(self, *args, **kwargs):
--> 63             return get_method(self)(*args, **kwargs)
    64
    65         wrapper.__module__ = owner.__module__

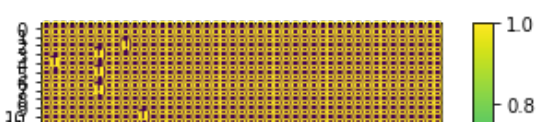
C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\cbook\deprecation.py in wrapper(*ar
gs, **kwargs)
    449             "parameter will become keyword-only %(removal)s.",
    450             name=name, obj_type=f"parameter of {func.__name__}()")
--> 451         return func(*args, **kwargs)
    452
    453         return wrapper

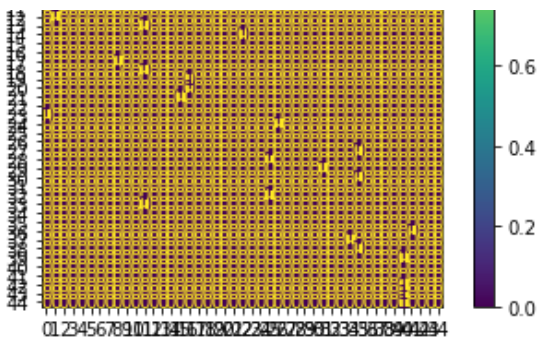
C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\axis.py in _set_ticklabels(self, la
bels, fontdict, minor, **kwargs)
   1791         if fontdict is not None:
   1792             kwargs.update(fontdict)
-> 1793         return self.set_ticklabels(labels, minor=minor, **kwargs)
   1794
   1795         @cbook._make_keyword_only("3.2", "minor")

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\axis.py in set_ticklabels(self, tic
klables, minor, **kwargs)
   1712             # remove all tick labels, so only error for > 0 ticklabels
   1713             if len(locator.locs) != len(ticklabels) and len(ticklabels) != 0:
-> 1714                 raise ValueError(
   1715                     "The number of FixedLocator locations"
   1716                     f" ({len(locator.locs)}), usually from a call to"

```

ValueError: The number of FixedLocator locations (45), usually from a call to `set_ticks`, does not match the number of ticklabels (29).





finalement, selon les résultats (score) obtenu sur chaque algorithmes : *SVM* : 0.8620689655172413 régression
 logistique : 0.8620689655172413 *Naïve bayes* : 0.8620689655172413 KNN : 0.8620689655172413 *arbres de*
décision : 0.8620689655172413 Multilayer Perceptron : 0.06896551724137931 *Adaboost* :
 0.7241379310344828 Random Forest : 0.8275862068965517

je recommande d'utiliser les 3 algorithmes : SVM ,DecisionTree ou Naïve Bayes qui donne le meilleur score 0.8620689655172413 .

Partie 3 : Regression lineaire

In [6]:

```
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.linear_model import LinearRegression
```

In [7]:

```
df=pd.read_excel('maison_bizertev2.xlsx')
```

In [8]:

```
df.head()
```

Out[8]:

	localisation	surface	nombre.chambre	prix
0	corniche	237	3	650
1	centre.ville	50	1	100
2	bhira	100	2	240
3	manzel.bourguiba	130	3	200
4	sejnen	170	3	180

In [9]:

```
X=df.iloc[:,1:3]
```

In [12]:

```
y=df.iloc[:,-1]
```

X prend les caracteristques
y cible prix

In [13]:

```
y
```

Out[13]:

```
0    650
1    100
2    240
3    200
4    180
5    570
6    177
7    120
8    133
9     83
10   600
11   100
12   300
13   110
14    58
15   140
16    90
17    80
18    79
```



```
array([589.82265716,  87.76906688, 228.33514079, 318.11467619,
        419.68775319, 279.12167929, 156.33089386,  75.07243226,
        100.01011000,  60.07570761,  80.08511070,  60.07570761])
```

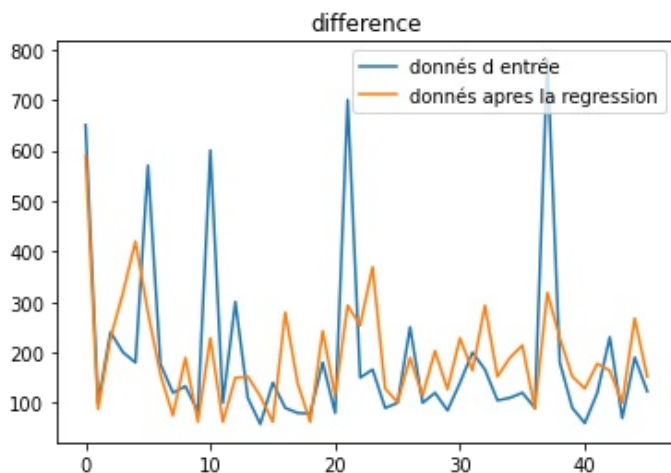
```
189.34214388, 62.37579764, 228.33514079, 62.37579764,
149.61600612, 152.15533304, 113.16233613, 62.37579764,
279.12167929, 138.55560538, 62.37579764, 241.93486845,
114.06542917, 292.72140694, 253.72841004, 368.90121469,
128.39829768, 101.36879454, 189.34214388, 114.06542917,
202.94187154, 126.76206379, 228.33514079, 163.94887463,
292.72140694, 152.15533304, 189.34214388, 214.00227227,
87.76906688, 318.11467619, 228.33514079, 152.15533304,
128.39829768, 177.54860229, 163.94887463, 100.46570151,
267.32813769, 152.15533304])
```

In [25]:

```
plt.plot(y,label='donnés d entrée')
plt.plot(regress.predict(X),label='donnés apres la regression')
plt.title('difference')
plt.legend(loc=0)
```

Out[25]:

<matplotlib.legend.Legend at 0xa5b598ae50>



la figure "difference", montre les données d'entrée en bleu et les données créer après la regression

In [27]:

```
mean_squared_error(y,regress.predict(X))
```

Out[27]:

19508.57823327489

In [28]:

```
r2_score(y,regress.predict(X))
```

Out[28]:

0.35342071621440785

Puisque le score est très faible, il existe deux solutions pour augmenter la performance de ce dernier:

- * ajouter plus des données d'entrée
- * ajouter des features

In [29]:

```
df2=pd.read_excel('maison_bizertev2.xlsx')
```

In [30]:

```
X=df2.iloc[:,1:3]
```

In [31]:

```
y=df2.iloc[:,-1]
```

In [33]:

```
y
```

Out[33]:

```
0      650000
1      100000
2      240000
3      200000
4      180000
5      570000
6      177000
7      120000
8      133000
9       83000
10     600000
11     100000
12     300000
13     110000
14      58000
15     140000
16      90000
17      80000
18      79000
19     180000
20      80000
21     700000
22     150000
23     166000
24      90000
25     100000
26     250000
27     100000
28     120000
29      85000
30     140000
31     200000
32     166000
33     105000
34     110000
35     120000
36      90000
37     780000
38     180000
39      90000
40      60000
41     120000
42     230000
43      70000
44     190000
45     123000
```

Name: prix, dtype: int64

In [34]:

```
regress=LinearRegression().fit(X,y)
```

In [35]:

```
from sklearn.metrics import mean_squared_error, r2_score
```

In [36]:

```
regress.coef_
```

Out[36]:

```
array([ 2539.32692493, 13599.72765721])
```

In [37]:

```
regress.intercept_
```

Out[37]:

```
-52797.00701924341
```

In [38]:

```
regress.predict(X)
```

Out[38]:

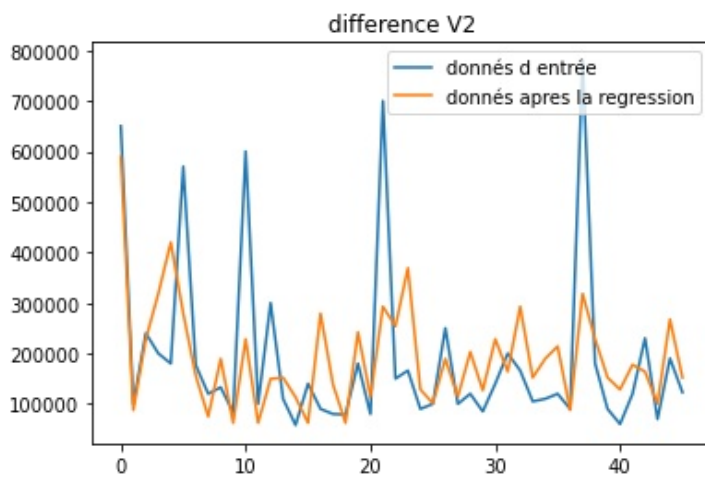
```
array([589822.65716126,  87769.06688456, 228335.14078837, 318114.67619354,
        419687.75319082, 279121.67928701, 156330.89385773,  75072.4322599 ,
        189342.14388184,  62375.79763524, 228335.14078837,  62375.79763524,
        149616.00611548, 152155.33304041, 113162.33613388,  62375.79763524,
        279121.67928701, 138555.6053832 ,  62375.79763524, 241934.86844558,
        114065.42916643, 292721.40694422, 253728.41003769, 368901.21469218,
        128398.29768348, 101368.79454177, 189342.14388184, 114065.42916643,
        202941.87153905, 126762.06379109, 228335.14078837, 163948.87463252,
        292721.40694422, 152155.33304041, 189342.14388184, 214002.27227133,
        87769.06688456, 318114.67619354, 228335.14078837, 152155.33304041,
        128398.29768348, 177548.60228973, 163948.87463252, 100465.70150922,
        267328.1376949 , 152155.33304041])
```

In [39]:

```
plt.plot(y,label='donnés d entrée')
plt.plot(regress.predict(X),label='donnés apres la regression')
plt.title('difference V2')
plt.legend(loc=0)
```

Out[39]:

```
<matplotlib.legend.Legend at 0xa5b4d80730>
```



In [40]:

```
mean_squared_error(y,regress.predict(X))
```

Out[40]:

```
19508578233.27489
```

In [41]:

```
r2_score(y,regress.predict(X))
```

Out[41]:

```
0.3534207162144076
```

Le même score que le premier donc je dois changer le fichier avec des autres données.

In [42]:

```
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.linear_model import LinearRegression
```

In [43]:

```
fichier2=pd.read_excel('maison_tunisie.xlsx')
```

In [44]:

```
fichier2.headad()
```

Out[44]:

	localisation	nombre de chambre	espace	prix
0	Hammamet	3	130	291340000
1	Manouba	1	95	132050000
2	Manouba	1	54	97730000
3	Manouba	2	98	161300000
4	Manouba	3	116	205010000

In [45]:

```
X=fichier2.iloc[:,1:3]
```

In [46]:

```
X
```

Out[46]:

	nombre de chambre	espace
0	3	130
1	1	95
2	1	54
3	2	98
4	3	116
5	4	147
6	1	71
7	2	90
8	3	110
9	2	77
10	3	94
11	1	77
12	2	92
13	3	119
14	1	71
15	3	137
16	2	90
17	3	107

	nombre de chambre	espace
18	2	91
19	3	108
20	2	76
21	3	104
22	1	65
23	2	95
24	3	116

In [47]:

```
y=fichier2.iloc[:,-1]
```

In [48]:

```
y
```

Out[48]:

```
0    291340000
1    132050000
2     97730000
3    161300000
4    205010000
5    343740000
6     84780000
7    108250000
8    135420000
9     86020000
10   105570000
11   143190000
12   170510000
13   214330000
14   139560000
15   259070000
16   124510000
17   148490000
18   140450000
19   165480000
20   103950000
21   142590000
22    96480000
23   142570000
24   174850000
```

Name: prix, dtype: int64

In [49]:

```
regression=LinearRegression().fit(X,y)
```

In [50]:

```
from sklearn.metrics import mean_squared_error,r2_score
```

In [51]:

```
regression.coef_
```

Out[51]:

```
array([-28151499.10115752,    3386328.13590612])
```

In [52]:

```
regression.intercept_
```

Out[52]:

```
-109402136.8234821
```

In [53]:

```
regression.predict(X)
```

Out[53]:

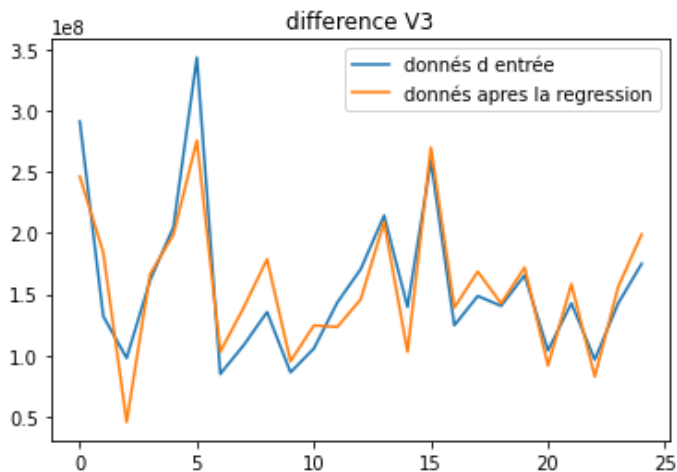
```
array([2.46366024e+08, 1.84147537e+08, 4.53080834e+07, 1.66155022e+08,
       1.98957430e+08, 2.75782103e+08, 1.02875662e+08, 1.39064397e+08,
       1.78639461e+08, 9.50421314e+07, 1.24458211e+08, 1.23193631e+08,
       1.45837053e+08, 2.09116414e+08, 1.02875662e+08, 2.70070320e+08,
       1.39064397e+08, 1.68480476e+08, 1.42450725e+08, 1.71866805e+08,
       9.16558033e+07, 1.58321492e+08, 8.25576929e+07, 1.55996038e+08,
       1.98957430e+08])
```

In [55]:

```
plt.plot(y,label='donnés d entrée')
plt.plot(regression.predict(X),label='donnés apres la regression')
plt.title('difference V3')
plt.legend(loc=0)
```

Out[55]:

<matplotlib.legend.Legend at 0xa5b5861910>



In [56]:

```
mean_squared_error(y,regression.predict(X))
```

Out[56]:

```
810363263799297.6
```

In [57]:

```
r2_score(y,regression.predict(X))
```

Out[57]:

```
0.7936630073931993
```

On peut constater que le score de ce fichier est très élevé par rapport aux scores précédents.